

Maximal δ -packings for finite VC-dimension set systems and application to discrepancy

Joint work with Victor-Emmanuel Brunel and Nabil Mustafa

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Table of Contents

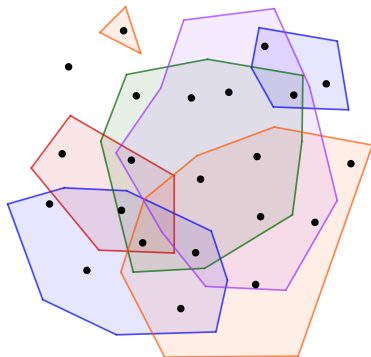
1 δ -packings

- Geometric set systems
- Combinatorial set systems

2 Algorithms for δ -packings/coverings

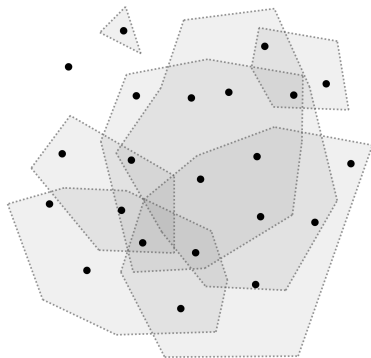
3 Application

- Discrepancy
- ε -approximation



Given (X, \mathcal{F}) a set system, a δ -packing is a collection $\mathcal{P} \subseteq \mathcal{F}$ such that:

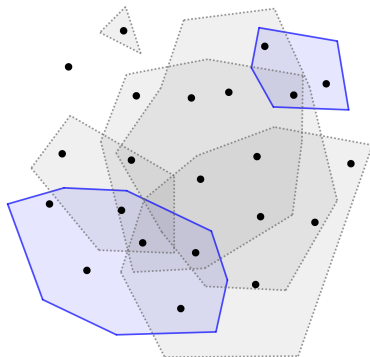
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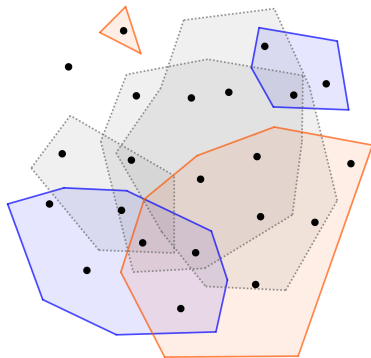
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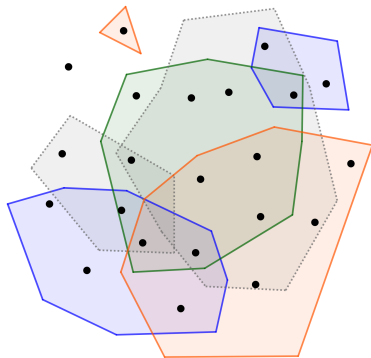
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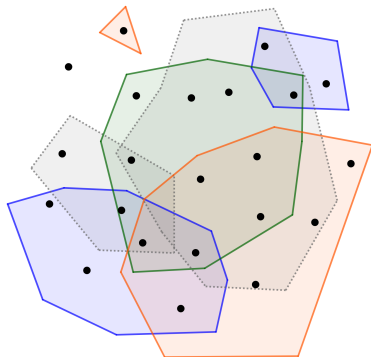
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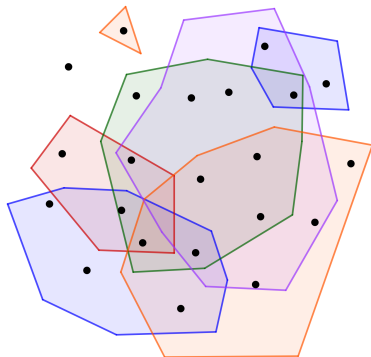
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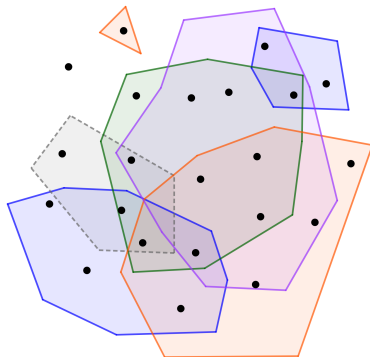
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We will study **maximal** packings



A δ -covering is a collection $\mathcal{C} \subseteq \mathcal{F}$ such that:

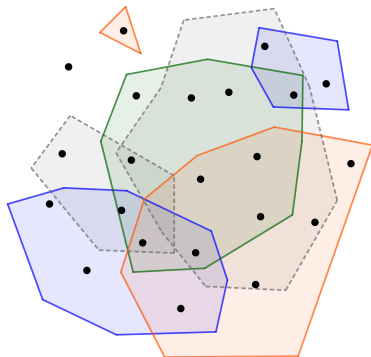
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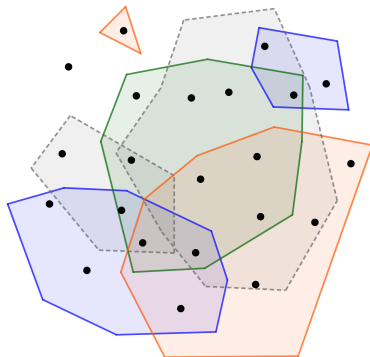
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A maximal δ -packing is a minimal δ -covering, however the opposite is not true

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Consider the set system $([n], 2^{[n]})$ and $\delta \in [0, n]$,

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- They form a packing of size $2^{\frac{n}{\delta}}$

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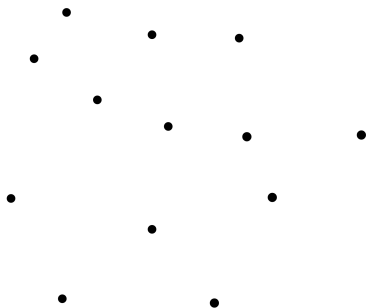
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 - [MWW93b] finds an approximate δ -covering in time $O\left(\frac{mn^2}{\delta^2}\right)$
 - A greedy algorithm finds a maximal δ -packing in time $O\left(mn\left(\frac{n}{\delta}\right)^d\right)$

(X, \mathcal{F}) induced by halfspaces

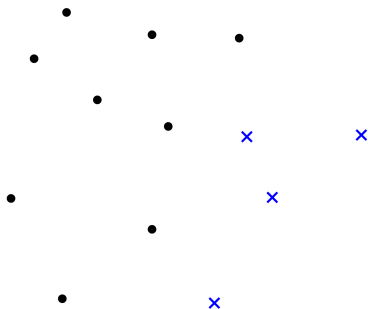
$E \subseteq X$ is in \mathcal{F} iff there exists a halfspace H such that $H \cap X = E$



Geometric set systems

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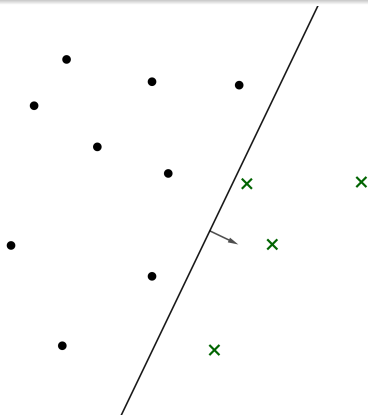
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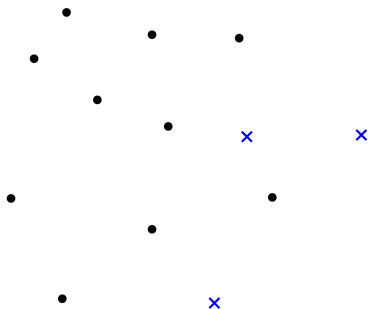
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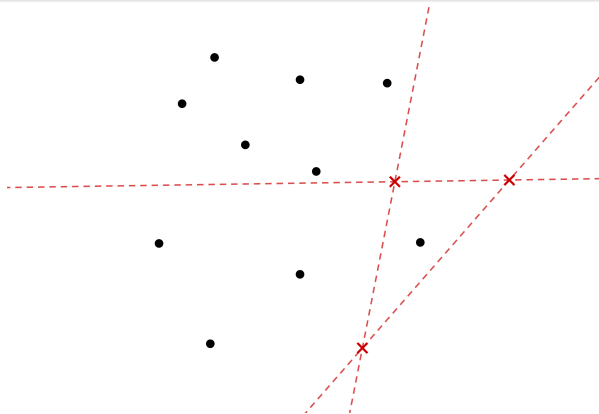
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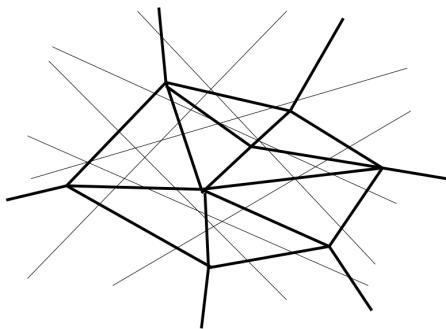
Theorem:

A packing of a set system induced by halfspaces in \mathbb{R}^d has size at most $O\left(\left(\frac{n}{\delta}\right)^d\right)$

Cuttings

We introduce a partition of space called *cuttings* that we will use for the proof of the packing bound.

Given a set H of hyperplanes in E^d , a $\frac{1}{r}$ -cutting for H is a collection of d -dimensional simplices with disjoint interiors, together covering E^d and such that the interior of each simplex intersects at most $\frac{|H|}{r}$ hyperplanes.



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Theorem [Cha93]

One can construct a $\frac{1}{r}$ -cutting of size $O(r^d)$ in time $O(nr^{d-1})$

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- The number of hyperplanes of \mathcal{G} intersecting the segment between two elements of Y is equal to the symmetric difference between the dual of these two elements in \mathcal{F}
- Thus a packing can only contain at most one element from each simplex which gives a bound of $O\left(\left(\frac{n}{\delta}\right)^d\right)$ on its size

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Sauer-Shelah lemma [Sau72; She72]

If (X, \mathcal{F}) has VC-dimension $\leq d$, then $|\mathcal{F}| = O(n^d)$

Packing lemma [Hau95]

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- This theorem shows that packing sets of VC-dimension d is similar to sphere packing in \mathbb{R}^d

Table of Contents

- 1 δ -packings
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- 2 Algorithms for δ -packings/coverings
- 3 Application
 - Discrepancy
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A *near-maximal* δ -packing of a set system (X, \mathcal{F}) is collection $\mathcal{P} \subseteq \mathcal{F}$ such that:

- \mathcal{P} is a δ -packing
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Theorem

We compute a near-maximal δ -packing in time $\tilde{O}\left(\frac{mn^2}{\delta^2} + \left(\frac{n}{\delta}\right)^{2d+2}\right)$.

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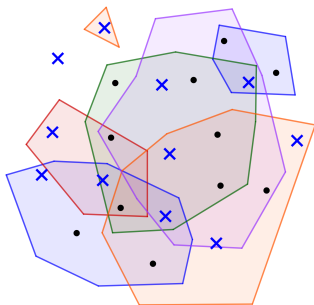
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$|X| = 22, \varepsilon = 1/22$ (no more than one element difference in each set)

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Algorithm Packing(\mathcal{F})

- 1: $A \leftarrow$ random sample of X of size $O\left(\frac{4n^2}{\delta^2}\right)$
 - 2: $\mathcal{P} \leftarrow \emptyset$
 - 3: **for** $F \in \mathcal{F}$ **do**
 - 4: **if** $\forall P \in \mathcal{P}, |\Delta(P \cap A, F \cap A)| \geq \frac{3n}{2\delta}$ **then**
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$$\begin{aligned} \rightarrow \delta &= \frac{3\delta}{2} - \frac{\delta}{2} \leq |\Delta(P, F)| \leq \\ &\frac{3\delta}{2} + \frac{\delta}{2} = 2\delta \\ \rightarrow \mathcal{P} &\text{ is a } \delta\text{-packing and a } 2\delta\text{-covering} \end{aligned}$$

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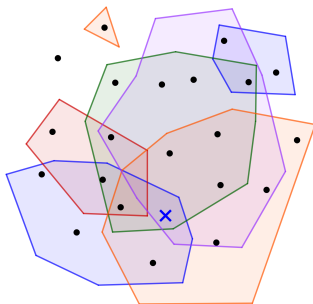
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$|X| = 22, \varepsilon = 5/22$ (sets with 5 elements and more)

Proof: An algorithm for small δ -covering [MWW93b]

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Algorithm δ -covering algorithm

```
1:  $\mathcal{S} \leftarrow \emptyset$ 
2:  $\mathcal{C} \leftarrow \emptyset$ 
3:  $N \leftarrow$  uniform random sample of
    $X$  of size  $O\left(\frac{n}{\delta} \log\left(\frac{n}{\delta}\right)\right)$ 
4: for  $F \in \mathcal{F}$  do
5:    $Q \leftarrow F \cap N$ 
6:   if  $Q \notin \mathcal{S}$  then
7:      $\mathcal{S} \leftarrow \mathcal{S} \cup \{Q\}$ 
8:      $\mathcal{C} \leftarrow \mathcal{C} \cup \{F\}$ 
9:   end if
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Proof: An algorithm for small δ -covering [MWW93b]

We show how to construct a small δ -covering:

Algorithm δ -covering algorithm

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3:  $N \leftarrow$  uniform random sample of
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4: for  $F \in \mathcal{F}$  do
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By Sauer-Shelah's lemma,
 $|\mathcal{C}| = O(|N|^d) = O\left(\left(\frac{n}{\delta}\right)^d \log^d\left(\frac{n}{\delta}\right)\right)$

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- This takes time $O\left(\left(\frac{n}{\delta}\right)^{2d+2} \log^d\left(\frac{n}{\delta}\right)\right)$ giving total runtime of $\tilde{O}\left(\frac{mn^2}{\delta^2} + \left(\frac{n}{\delta}\right)^{2d+2}\right)$

Table of Contents

1 δ -packings

- Geometric set systems
- Combinatorial set systems

2 Algorithms for δ -packings/coverings

3 Application

- Discrepancy
- ε -approximation

Set discrepancy

We want to compute a 2-coloring $\chi : X \rightarrow \{-1, 1\}$ s.t.

$\forall F \in \mathcal{F}, \chi(F) = \sum_{x \in F} \chi(x)$ is small. We call discrepancy and denote:

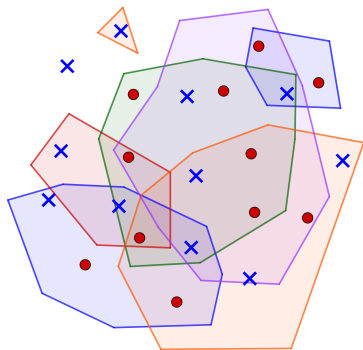
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Theorem:

Given a set system of VC-dimension $\leq d$, we compute a coloring with discrepancy $O\left(n^{1/2-1/2d}\right)$ in time $O\left(mn^{2/d} + n^3 \log^{3d}(n)\right)$

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- This gives $\text{disc}(F) \leq \text{disc}(P) + \text{disc}(P \setminus F) + \text{disc}(F \setminus P) = O\left(\sqrt{n^{1-1/d} \log(m)}\right)$

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- We use [LM12] with n points and $\sum_{i=0}^{\log \log m} |\mathcal{P}_i| = n \log(n)$ sets which gives runtime $O(n^3 \log^{3d}(n))$

Results on ε -approximation

- By computing iterated discrepancy, we obtain an ε -approximation of size $O\left(\varepsilon^{-\frac{2d}{d+1}}\right)$ in time $O\left(\left[mn^{2/d} + n^3 \log^{3d}(n)\right] \log\left(\frac{1}{\varepsilon}\right)\right)$

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- Using the *merge and reduce* framework [CM96], this gives an ε -approximation of size $O\left(\varepsilon^{-\frac{2d}{d+1}}\right)$ in time $\tilde{O}\left(\frac{n}{\varepsilon^{2d}}\right)$

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Open problems

- Is it possible to find a polynomial-time algorithm for (near-)maximal packings with degree $< d$ (for δ bigger than a constant)?
- Is it possible to find faster algorithm set systems with smaller packing size than Haussler's bound?

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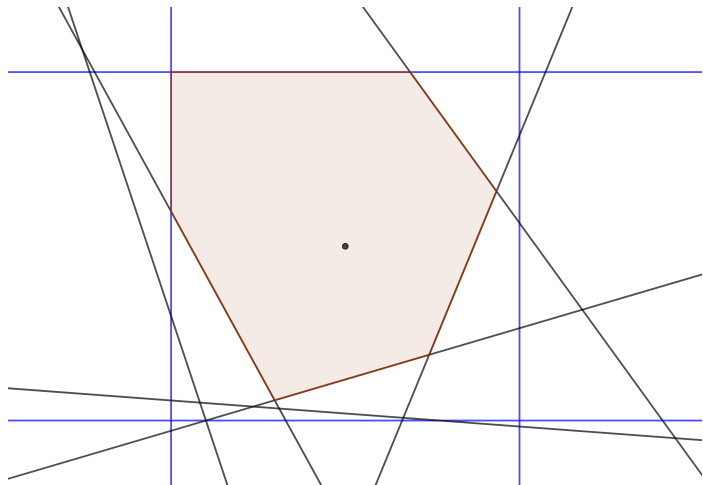
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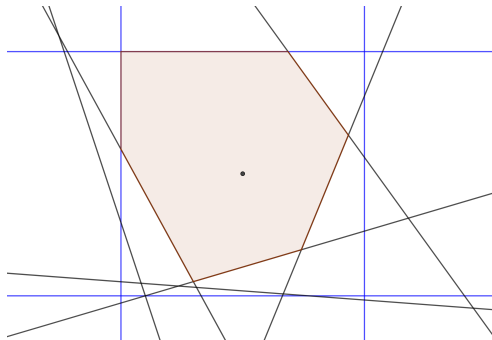
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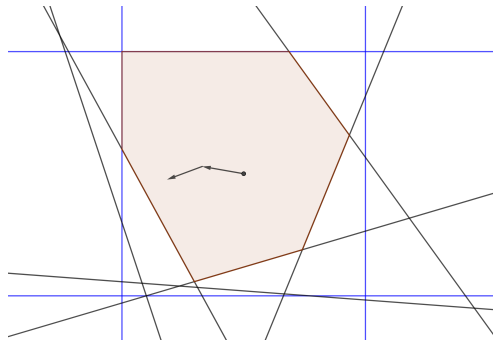
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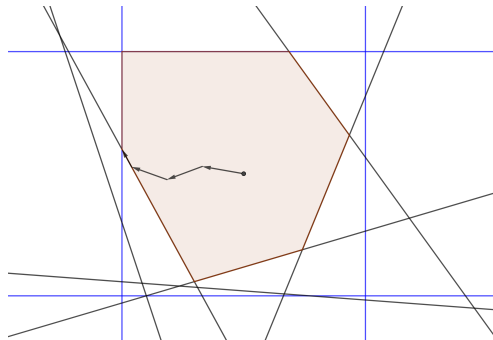
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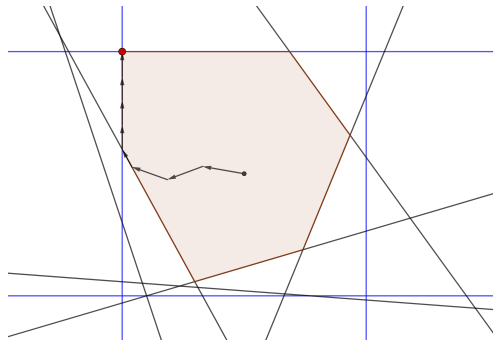
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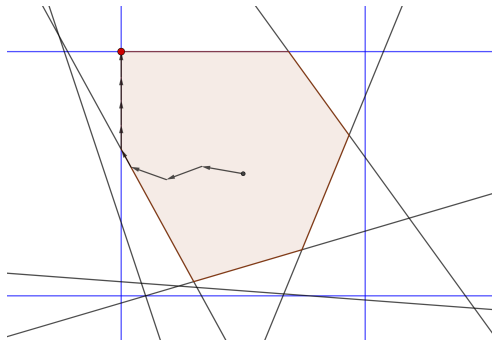
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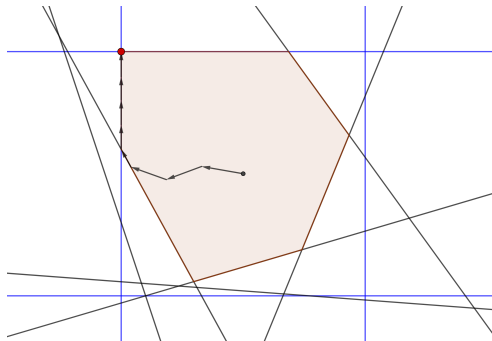
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Entropy condition

$$\sum_{F \in \mathcal{F}} e^{-\lambda_F^2} \leq n$$